

Meta-Analysis of the Effects of Voting Advice Applications

Online Appendix

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APPENDIX A OVERVIEW OF COLLECTED STUDIES

FIGURE A1: Geographic distribution of VAAs. The purple filling indicates countries covered by studies that were included in the meta-analysis. *Source:* 2016 Global Census of the ECPR Research Network on Voting Advice Applications (http://vaa-research.net/?page_id=146).

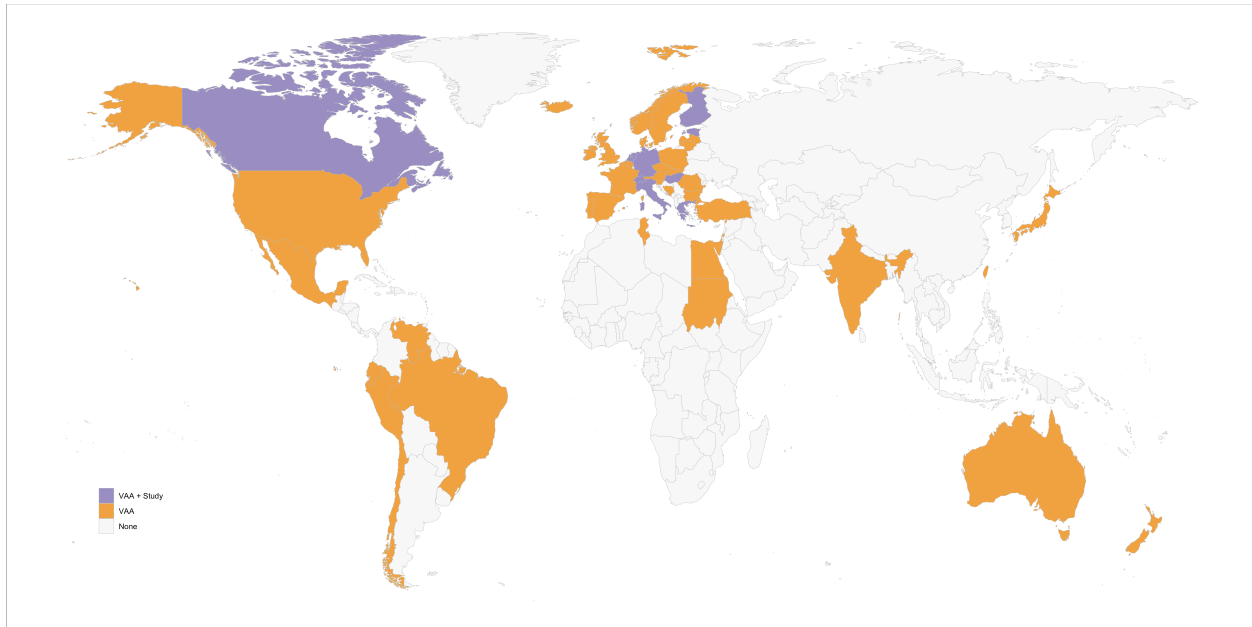


FIGURE A2: Publication timeline of 22 studies included in the meta-analysis, by study type.

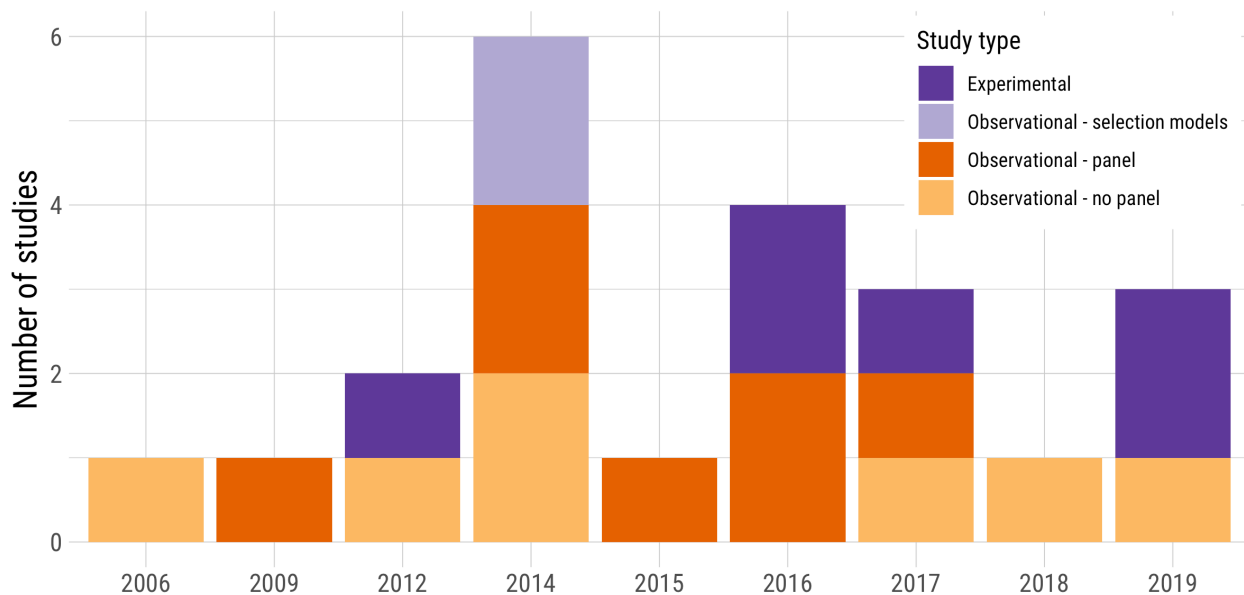


TABLE A1: Overview of selected within-study effects excluded from the meta-analysis.

Study	Election	Outcome of interest	Exclusion reason
Alvarez, Levin, Trechsel and Vassil (2014)	EU 2009	Perceived utility	IVAR; NOV
Alvarez, Levin, Mair and Trechsel (2014)	EU 2009	Vote choice	IVAR
Boogers (2006)	Netherlands 2006	Self-assessed impact on information seeking	IVAR; NOV
Boogers (2006)	Netherlands 2006	Turnout	IVAR
Boogers (2006)	Netherlands 2006	Vote choice	IVAR
De Rosa (2010)	Italy - EU 2009	Motivation to seek information	IVAR; NOV
De Rosa (2010)	Italy - EU 2009	Issue knowledge	IVAR
De Rosa (2010)	Italy - EU 2009	Turnout	IVAR
Dinas et al. (2014)	EU 2009	Turnout	IVAR
Enyedi (2016)	Hungary 2010	Vote choice	IVAR
Fivaz and Nadig (2010)	Switzerland 2007	Turnout	NMB; IMOV
Fivaz and Nadig (2010)	Switzerland 2007	Vote choice	NMB; IMOV
Garry et al. (2018)	N. Ireland 2016	Party support (propensity to vote for)	IMOV
Garzia et al. (2017)	Finland 2007	Turnout	NISP - Duplicate of Garzia and Angelis (2014)
Garzia et al. (2017)	Finland 2011	Turnout	NISP - Duplicate of Garzia and Angelis (2014)
Garzia et al. (2017)	Germany 2009	Turnout	NISP - Duplicate of Garzia and Angelis (2014)
Garzia et al. (2017)	Netherlands 2003	Turnout	NISP - Duplicate of Garzia and Angelis (2014)
Garzia et al. (2017)	Netherlands 2006	Turnout	NISP - Duplicate of Garzia and Angelis (2014)
Garzia et al. (2017)	Netherlands 2010	Turnout	NISP - Duplicate of Garzia and Angelis (2014)
Garzia et al. (2017)	Switzerland 2007	Turnout	NISP - Duplicate of Garzia and Angelis (2014)
Garzia et al. (2017)	Switzerland 2011	Turnout	NISP - Duplicate of Garzia and Angelis (2014)
Gemenis (2018)	Greece 2015	Turnout	IVAR
Israel et al. (2016)	Germany - EU 2014	Vote choice	IVAR
Israel et al. (2017)	Germany - EU 2014	Vote choice	IVAR
Israel et al. (2017)	Germany - EU 2014	Vote choice	IVAR
Kamoen et al. (2015)	Netherlands 2012	Issue knowledge	IVAR
Kamoen et al. (2015)	Netherlands 2012	Vote choice	IVAR
Ladner and Pianzola (2010)	Switzerland 2007	Turnout	NMB
Ladner et al. (2012)	Switzerland 2007	Vote choice	IVAR
Mahéo (2017)	Canada 2014	Attention to campaign	NOV
Mahéo (2017)	Canada 2014	Information seeking	NOV
Manavopoulos et al. (2018)	Germany 2017	Information seeking	IVAR; NOV
Marschall and Schmidt (2008)	Germany 2005	Information seeking	NOV
Marschall and Schmidt (2008)	Germany 2005	Turnout	NMB
Marschall and Schmidt (2010)	Germany - EU 2009	Turnout	NMB
Marschall and Schultze (2012 <i>b</i>)	Germany 2009	Turnout	NISP - Duplicate of Marschall and Schultze (2012 <i>a</i>)
Nuytemans et al. (2010)	Belgium 2009	Vote choice	NSI
Pianzola et al. (2019)	Switzerland 2011	Propensity to vote for most preferred party	IMOV
Ramos et al. (2019)	EU 2014	Turnout	NMB
Ruusuvirta and Rosema (2009)	Netherlands 2006	Vote choice	NSI
Schultze (2013)	Germany 2009	Issue knowledge	NISP - Duplicate of Schultze (2014)
Walgrave et al. (2008)	Belgium 2004	Vote choice	NSI
Wall et al. (2014)	Netherlands 2010	Vote choice	NMB; IMOV
Wang (2016)	Taiwan 2012	Turnout	IVAR
Westle et al. (2015)	Germany 2013	Issue knowledge	NSI

Notes: IVAR: VAA usage not independent variable; NOV: No measure of turnout, vote choice, issue knowledge as outcome; NMB: No meaningful baselines; NSI: Insufficient information about modeling; IMOV: Incompatible measurement of outcome variable; NISP: Not independent separate publication

TABLE A2: Overview of VAA turnout effects studies used.

Study	Election	Sample size	VAA users	Model	Effect (orig)	SE (orig)	Effect (conv)	SE (conv)	Study Design Type
Enyedi (2016)	Hungary 2010	1727	1164	mean differences	3.10		0.29	0.16	Experimental
Garzia and Angelis (2014)	Finland 2007	1062	313	logit	0.53	0.15	0.53	0.15	Obs — no panel
Garzia and Angelis (2014)	Finland 2011	1030	443	logit	0.24	0.15	0.24	0.15	Obs — no panel
Garzia and Angelis (2014)	Germany 2009	1187	116	logit	0.15	0.17	0.15	0.17	Obs — no panel
Garzia and Angelis (2014)	Netherlands 2003	1016	325	logit	0.65	0.27	0.65	0.27	Obs — no panel
Garzia and Angelis (2014)	Netherlands 2006	1793	687	logit	0.46	0.15	0.46	0.15	Obs — no panel
Garzia and Angelis (2014)	Netherlands 2010	1693	706	logit	0.59	0.14	0.59	0.14	Obs — no panel
Garzia and Angelis (2014)	Switzerland 2007	3127	256	logit	0.26	0.07	0.26	0.07	Obs — no panel
Garzia and Angelis (2014)	Switzerland 2011	3254	368	logit	0.32	0.07	0.32	0.07	Obs — no panel
Garzia et al. (2017)	Italy 2013	888	454	mean differences	10.70		0.54	0.15	Experimental
Garzia et al. (2017)	EU 2009	19592	921	logit	1.11	0.11	1.11	0.11	Obs — no panel
Garzia et al. (2017)	Finland 2003	1152	255	logit	1.17	0.31	1.17	0.31	Obs — no panel
Garzia et al. (2017)	Germany 2013	1820	260	logit	1.08	0.33	1.08	0.33	Obs — no panel
Garzia et al. (2017)	Netherlands 2012	1574	631	logit	1.84	0.27	1.84	0.27	Obs — no panel
Gemenis and Rosema (2014)	Netherlands 2006	2356	903	logit	1.44	0.25	1.44	0.25	Obs — panel
Gemenis (2018)	Greece 5/2012	890	NA	entropy bal + LR	1.01	0.59	1.01	0.59	Obs — selection/matching
Gemenis (2018)	Greece 6/2012	896	NA	entropy bal + LR	0.97	0.64	0.97	0.64	Obs — selection/matching
Gemenis (2018)	Greece 1/2015	1013	NA	entropy bal + LR	0.09	0.50	0.09	0.50	Obs — selection/matching
Gemenis (2018)	Greece 9/2015	806	NA	entropy bal + LR	0.03	0.43	0.03	0.43	Obs — selection/matching
Germann and Gemenis (2019)	Switzerland 2007	3578	322	entropy bal + LR	0.60	0.19	0.60	0.19	Obs — selection/matching
Germann and Gemenis (2019)	Switzerland 2011	3657	366	entropy bal + LR	0.67	0.18	0.67	0.18	Obs — selection/matching
Germann and Gemenis (2019)	Switzerland 2015	2931	410	entropy bal + LR	0.64	0.17	0.64	0.17	Obs — selection/matching
Heinsohn et al. (2016)	EU - Germany - EU 2014	449	127	log panel reg with FEs	0.35	0.09	0.35	0.09	Obs — panel
Mahéo (2017)	Canada 2014	277	138	logit	0.40	0.40	0.40	0.40	Experimental
Marschall and Schultze (2012c)	Germany 2009	1153	421	logit	0.60	0.25	0.60	0.25	Obs — no panel
Munzert et al. (Forthcoming)	Germany 2017	979	499	IV model	-0.01	0.05	-0.04	0.20	Experimental
Mykkänen and Moring (2006)	EU - Finland - EU 2004	1362	NA	logit	0.63	0.23	0.63	0.23	Obs — no panel
Mykkänen and Moring (2006)	Finland 2003	1511	NA	logit	1.01	0.23	1.01	0.23	Obs — no panel
Ruusuvirta and Rosema (2009)	Netherlands 2006	2356	892	frequency tables	6.00		1.21	0.22	Obs — panel
Vassil (2011)	EU - Estonia - EU 2009	279	97	IV model	0.13	0.06	0.52	0.24	Experimental

TABLE A3: Overview of VAA vote choice effects studies used.

Study	Election	Sample size	VAA users	Model	Effect (orig)	SE (orig)	Effect (conv)	SE (conv)	Study Design Type
Andreadis and Wall (2014)	Finland 2003	578	153	probit	0.43	0.14	0.68	0.23	Obs — panel
Andreadis and Wall (2014)	Finland 2007	617	188	probit	0.32	0.14	0.52	0.22	Obs — panel
Andreadis and Wall (2014)	Finland 2011	526	237	probit	0.20	0.14	0.32	0.22	Obs — panel
Andreadis and Wall (2014)	Germany 2009	1175	147	probit	0.25	0.12	0.40	0.20	Obs — panel
Andreadis and Wall (2014)	Netherlands 2003	1170	387	probit	0.14	0.09	0.22	0.14	Obs — panel
Andreadis and Wall (2014)	Netherlands 2006	1721	685	probit	0.26	0.07	0.42	0.12	Obs — panel
Andreadis and Wall (2014)	Netherlands 2010	1537	661	probit	0.24	0.07	0.39	0.12	Obs — panel
Andreadis and Wall (2014)	Switzerland 2007	1787	166	probit	0.10	0.12	0.16	0.19	Obs — panel
Andreadis and Wall (2014)	Switzerland 2011	1645	206	probit	0.20	0.10	0.32	0.17	Obs — panel
Enyedi (2016)	Hungary 2010	1727	1124	mean differences			0.05	0.12	Experimental
Klein Kranenburg (2015)	Netherlands 2006	1628	624	logit	0.29	0.13	0.29	0.13	Obs — panel
Klein Kranenburg (2015)	Netherlands 2010	1485	619	logit	0.41	0.16	0.41	0.16	Obs — panel
Klein Kranenburg (2015)	Netherlands 2012	1026	411	logit	0.47	0.18	0.47	0.18	Obs — panel
Klein Kranenburg (2015)	Netherlands 2015	5211	1516	logit	0.48	0.07	0.48	0.07	Obs — panel
Kleinnijenhuis et al. (2019)	Netherlands 2010	1159	426	logit with REs	1.34	0.52	1.34	0.52	Obs — panel
Mahéo (2016)	Canada 2014	211	211	linear regression	-0.24	0.30	-0.24	0.30	Experimental
Munzert et al. (Forthcoming)	Germany 2017	923	483	IV model	0.05	0.07	0.20	0.28	Experimental
Pianzola (2014a)	Switzerland 2007	4067	NA	IV model	0.18	0.04	0.72	0.16	Obs — selection/matching
Pianzola (2014b)	Switzerland 2011	9163	NA	Heckman + PS matching	0.16	0.02	0.66	0.07	Obs — selection/matching
Pianzola et al. (2019)	Switzerland 2011	1775	1349	IV model	-0.07	0.08	-0.28	0.32	Experimental
Vassil (2011)	EU - Estonia - EU 2009	394	186	IV model	0.15	0.10	0.60	0.40	Experimental

TABLE A4: Overview of VAA issue knowledge effects studies used.

Study	Election	Sample size	VAA users	Model	Effect (orig)	SE (orig)	Effect (conv)	SE (conv)	Study Design Type
Heinsohn et al. (2016)	EU - Germany - EU 2014	1569	NA	panel regression with FEs	0.23	0.07			Obs — panel
Munzert et al. (Forthcoming)	Germany 2017	979	499	IV model	0.04	0.01	0.16	0.04	Experimental
Schultze (2014)	Germany 2009	1145	440	SEM probit	0.34	0.05	0.55	0.08	Obs — no panel
van de Pol (2016)	Netherlands 2014	5571	3522	entropy bal + LR	-0.06	0.02	-0.06	0.02	Obs — selection/matching

APPENDIX B SUPPORTING TABLES AND FIGURES

TABLE B1: Turnout study design results of mixed effects moderator analysis

	Estimate	k	Std. Error	Z. Value	Conf. Low	Conf. High	P. Value
Experimental	0.21	5	0.17	1.29	-0.11	0.54	0.2
Observational - no panel	0.8***	15	0.16	4.94	0.48	1.12	< 0.001
Observational - panel	0.91***	3	0.23	3.93	0.46	1.36	< 0.001
Observational - selection/matching	0.56*	7	0.27	2.04	0.02	1.1	0.04

Notes: The estimates are presented in log-odds. The tests for residual heterogeneity and of moderators are significant at the 95%: QE(df = 26) = 133.0338, p-val < .0001, QM(df = 4) = 44.9021, p-val < .0001

TABLE B2: Vote choice study design results of mixed effects moderator analysis

	Estimate	k	Std. Error	Z. Value	Conf. Low	Conf. High	P. Value
Experimental	0.04	5	0.09	0.45	-0.14	0.23	0.65
Observational - panel	0.4***	14	0.04	10.37	0.33	0.48	< 0.001
Observational - selection/matching	0.67***	2	0.07	10.16	0.54	0.8	< 0.001

Notes: The estimates are presented in log-odds. The test for residual heterogeneity is not significant at the 95%: QE(df = 18) = 15.0816, p-val = 0.6564. The test of moderators is significant: QM(df = 3) = 210.7735, p-val < .0001

TABLE B3: Turnout country results of mixed effects moderator analysis

	Estimate	k	Std. Error	Z. Value	Conf. Low	Conf. High	P. Value
Finland	0.81***	4	0.16	5.16	0.5	1.12	< 0.001
Germany	0.48***	4	0.17	2.82	0.15	0.82	< 0.001
Netherlands	1.07***	6	0.15	7.23	0.78	1.36	< 0.001
Other	0.43***	11	0.12	3.49	0.19	0.67	< 0.001
Switzerland	0.71***	5	0.15	4.79	0.42	1.01	< 0.001

Notes: The estimates are presented in log-odds. The tests for residual heterogeneity and of moderators are significant at the 95%: QE(df = 25) = 102.3942, p-val < .0001, QM(df = 5) = 58.2662, p-val < .0001

TABLE B4: Vote choice country results of mixed effects moderator analysis

	Estimate	k	Std. Error	Z. Value	Conf. Low	Conf. High	P. Value
Finland	0.63***	3	0.18	3.57	0.29	0.98	< 0.001
Germany	0.45*	2	0.2	2.29	0.06	0.84	0.02
Netherlands	0.5***	8	0.13	3.83	0.24	0.75	< 0.001
Other	0.07	3	0.19	0.37	-0.31	0.45	0.71
Switzerland	0.43***	5	0.13	3.41	0.18	0.68	< 0.001

Notes: The estimates are presented in log-odds. The tests for residual heterogeneity and of moderators are significant at the 95%: QE(df = 16) = 27.5137, p-val = 0.0361, QM(df = 5) = 19.6308, p-val = 0.0015

TABLE B5: Turnout election year results of mixed effects moderator analysis

	Estimate	k	Std. Error	Z. Value	Conf. Low	Conf. High	P. Value
2003-2005	0.86***	4	0.23	3.73	0.41	1.32	< 0.001
2006-2008	0.75***	6	0.19	4.02	0.39	1.12	< 0.001
2009-2011	0.63***	9	0.14	4.51	0.36	0.91	< 0.001
2012-2014	0.68***	7	0.18	3.68	0.32	1.04	< 0.001
2015-2017	0.27	4	0.22	1.23	-0.16	0.7	0.22

Notes: The estimates are presented in log-odds. The tests for residual heterogeneity and of moderators are significant at the 95%: QE(df = 25) = 127.1568, p-val < .0001, QM(df = 5) = 41.8469, p-val < .0001

TABLE B6: Vote choice election year results of mixed effects moderator analysis

	Estimate	k	Std. Error	Z. Value	Conf. Low	Conf. High	P. Value
2003-2005	0.33	2	0.18	1.86	-0.02	0.67	0.06
2006-2008	0.34**	5	0.13	2.55	0.08	0.61	0.01
2009-2011	0.35***	10	0.12	2.88	0.11	0.59	< 0.001
2012-2014	0.33	2	0.2	1.64	-0.06	0.72	0.1
2015-2017	0.44***	2	0.15	2.98	0.15	0.73	< 0.001

Notes: The estimates are presented in log-odds. The tests for residual heterogeneity is significant at the 95%: QE(df = 16) = 44.0260, p-val = 0.0002. The test of moderators is not significant: QM(df = 5) = 11.0442, p-val = 0.0505

TABLE B7: Turnout election type results of mixed effects moderator analysis

	Estimate	k	Std. Error	Z. Value	Conf. Low	Conf. High	P. Value
First order	0.65***	25	0.12	5.33	0.41	0.89	< 0.001
Second order	0.57***	5	0.18	3.26	0.23	0.91	< 0.001

Notes: The estimates are presented in log-odds. The tests for residual heterogeneity and of moderators are significant at the 95%: QE(df = 28) = 131.7178, p-val < .0001, QM(df = 2) = 32.4878, p-val < .0001

TABLE B8: Vote choice election type results of mixed effects moderator analysis

	Estimate	k	Std. Error	Z. Value	Conf. Low	Conf. High	P. Value
First order	0.35***	18	0.12	2.98	0.12	0.57	< 0.001
Second order	0.43***	3	0.14	3	0.15	0.7	< 0.001

Notes: The estimates are presented in log-odds. The tests for residual heterogeneity and of moderators are significant at the 95%: QE(df = 19) = 45.5668, p-val = 0.0006, QM(df = 2) = 10.7327, p-val = 0.0047

TABLE B9: Turnout sample size results of mixed effects moderator analysis

	Estimate	k	Std. Error	Z. Value	Conf. Low	Conf. High	P. Value
Large (>3000)	0.75***	8	0.12	6.09	0.51	0.99	< 0.001
Medium (2000-3000)	0.84***	14	0.13	6.75	0.6	1.09	< 0.001
Small (<1000)	0.27	8	0.14	1.91	-0.01	0.55	0.06

Notes: The estimates are presented in log-odds. The tests for residual heterogeneity and of moderators are significant at the 95%: QE(df = 27) = 133.6281, p-val < .0001, QM(df = 3) = 48.1064, p-val < .0001

TABLE B10: Vote choice sample size results of mixed effects moderator analysis

	Estimate	k	Std. Error	Z. Value	Conf. Low	Conf. High	P. Value
Large (>3000)	0.52***	3	0.09	5.87	0.35	0.7	< 0.001
Medium (2000-3000)	0.28***	12	0.08	3.4	0.12	0.44	< 0.001
Small (<1000)	0.35**	6	0.12	2.79	0.1	0.59	0.01

Notes: The estimates are presented in log-odds. The test for residual heterogeneity is not significant at the 95%: QE(df = 18) = 26.6059, p-val = 0.0867. The test of moderators is significant: QM(df = 3) = 38.3041, p-val < .0001

TABLE B11: Turnout sampling type results of mixed effects moderator analysis

	Estimate	k	Std. Error	Z. Value	Conf. Low	Conf. High	P. Value
Non-probability	0.56***	11	0.14	4.08	0.29	0.83	< 0.001
Probability	0.73***	16	0.14	5.2	0.46	1.01	< 0.001

Notes: The estimates are presented in log-odds. The tests for residual heterogeneity and of moderators are significant at the 95%: QE(df = 25) = 133.5, p-val < .0001, QM(df = 2) = 27.99, p-val < .0001

TABLE B12: Vote choice sample type results of mixed effects moderator analysis

	Estimate	k	Std. Error	Z. Value	Conf. Low	Conf. High	P. Value
Convenience	-0.26	2	0.28	-0.94	-0.8	0.28	0.35
Non-probability	0.5***	9	0.11	4.45	0.28	0.72	< 0.001
Probability	0.42***	10	0.11	3.77	0.2	0.64	< 0.001

Notes: The estimates are presented in log-odds. The tests for residual heterogeneity and of moderators are significant at the 95%: QE(df = 18) = 34.62, p-val = 0.011, QM(df = 3) = 21.14, p-val < .0001

TABLE B13: Operationalization of vote choice results of mixed effects moderator analysis

	Estimate	k	Std. Error	Z. Value	Conf. Low	Conf. High	P. Value
Between elections	0.46***	16	0.11	4.24	0.25	0.67	< 0.001
In-campaign	0.19	5	0.16	1.23	-0.11	0.5	0.22

Notes: The estimates are presented in log-odds. The tests for residual heterogeneity and of moderators are significant at the 95%: QE(df = 19) = 34.43, p-val = 0.016, QM(df = 3) = 19.45, p-val < .0001

FIGURE B1: Mixed effects moderator analysis on country of election

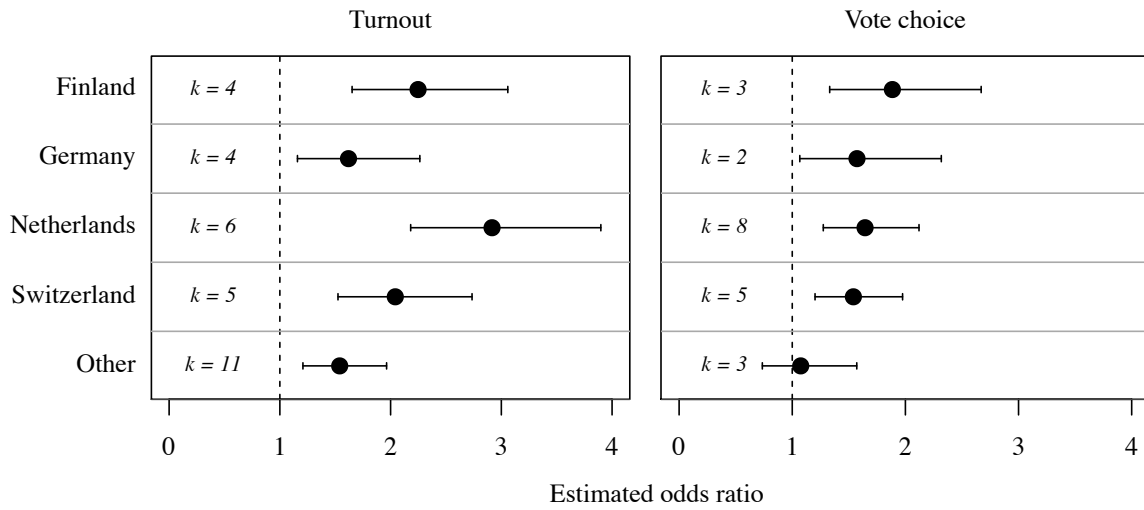


FIGURE B2: Mixed effects moderator analysis on election year

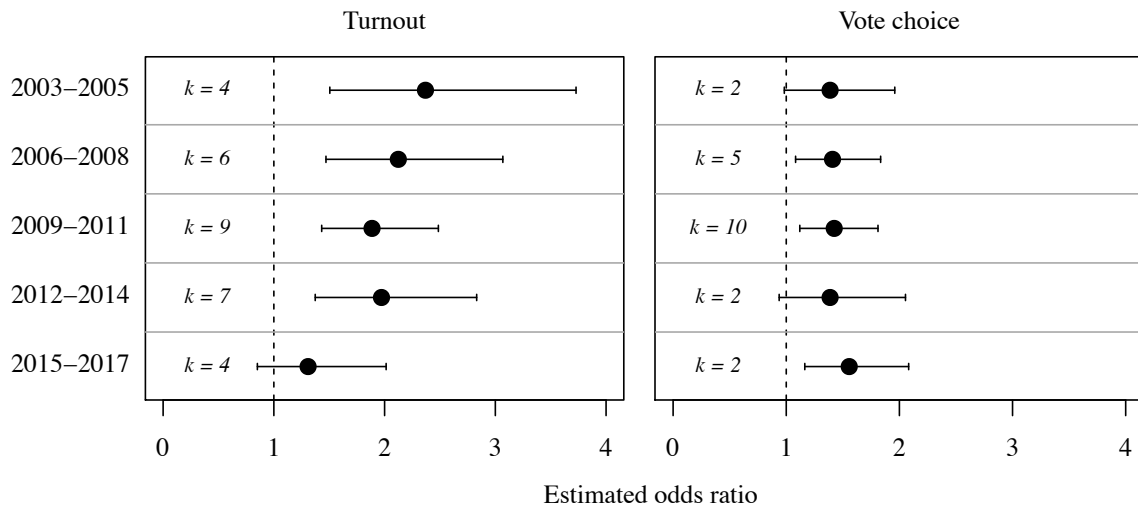


FIGURE B3: Mixed effects moderator analysis on election type

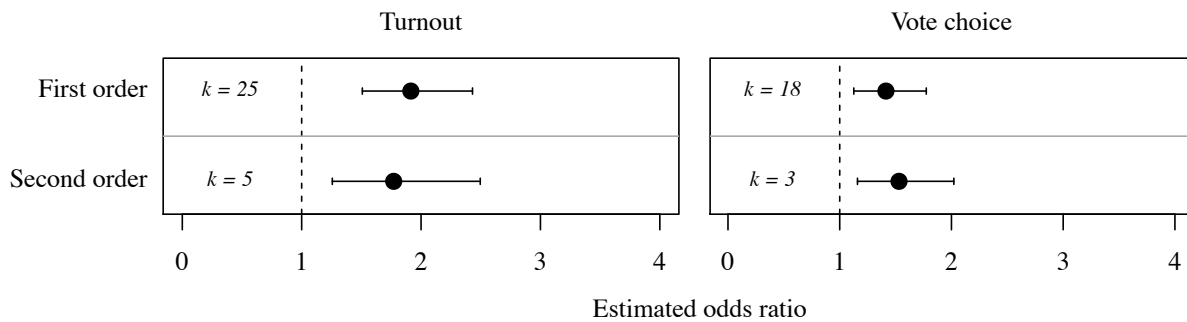


FIGURE B4: Mixed effects moderator analysis on sample size

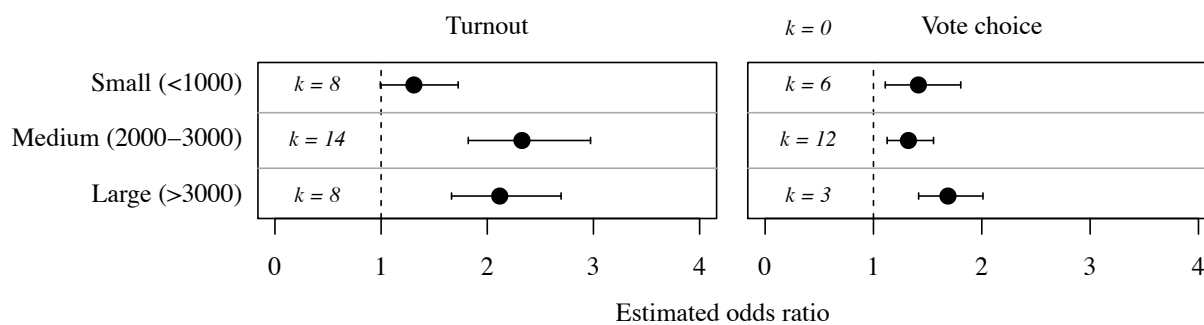


FIGURE B5: Mixed effects moderator analysis on sample type

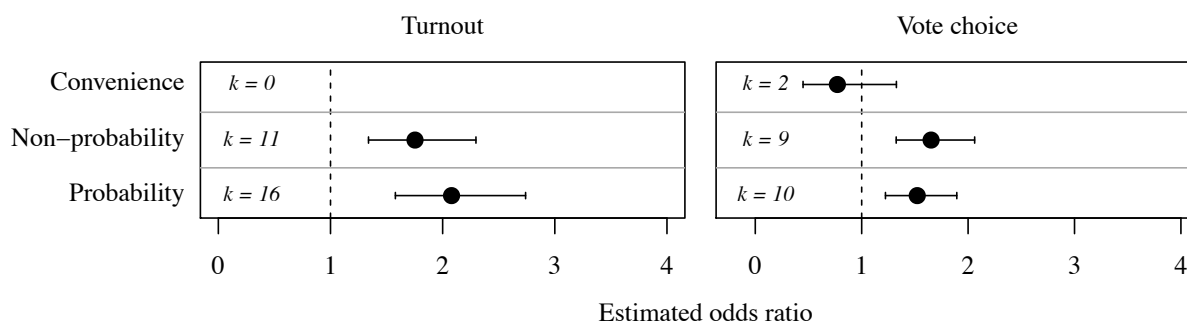


FIGURE B6: Mixed effects moderator analysis on operationalization of vote choice

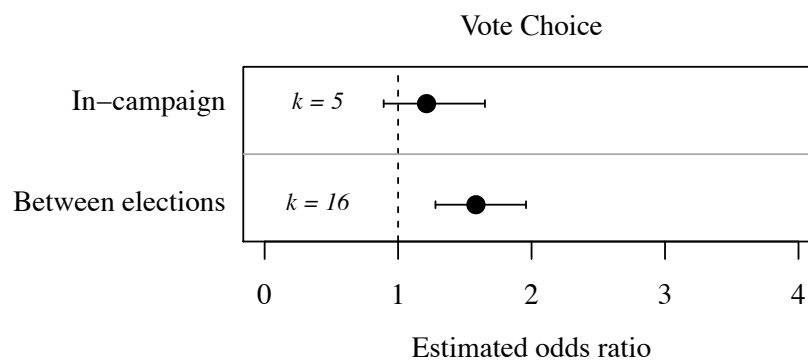


FIGURE B7: Mixed effects moderator analysis, multiple moderators

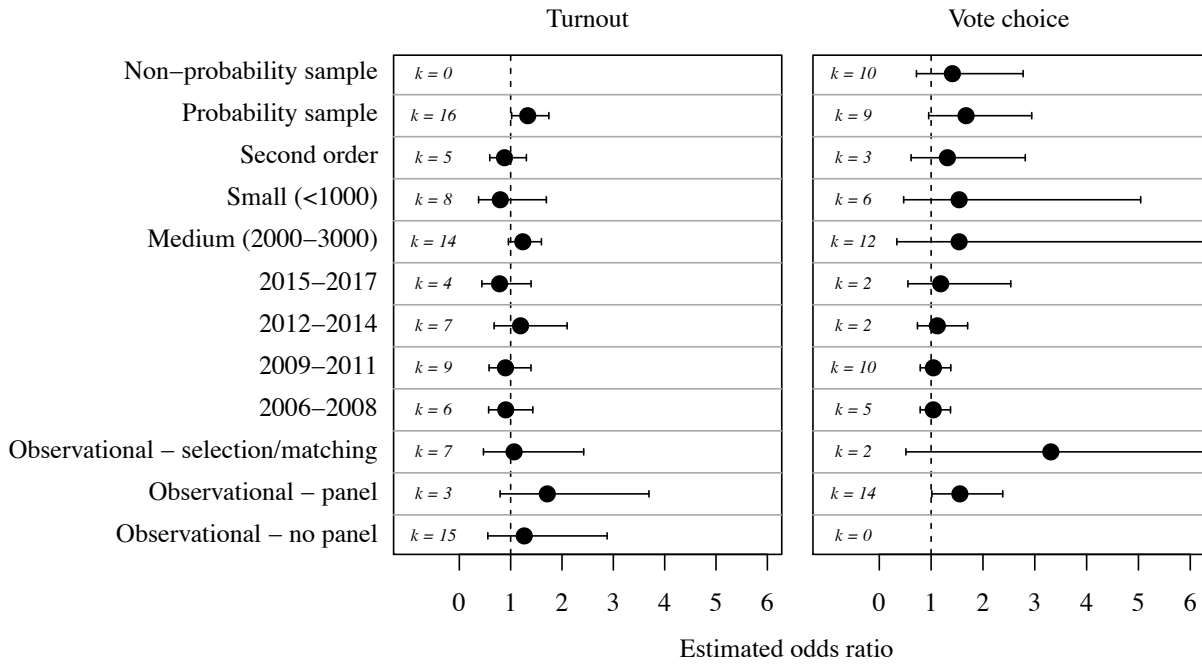
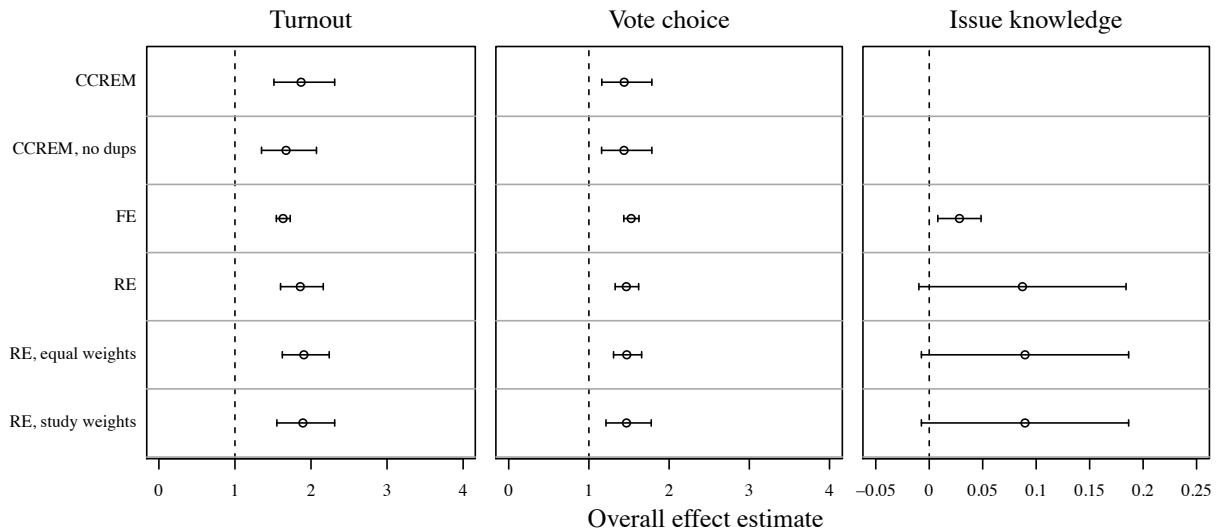
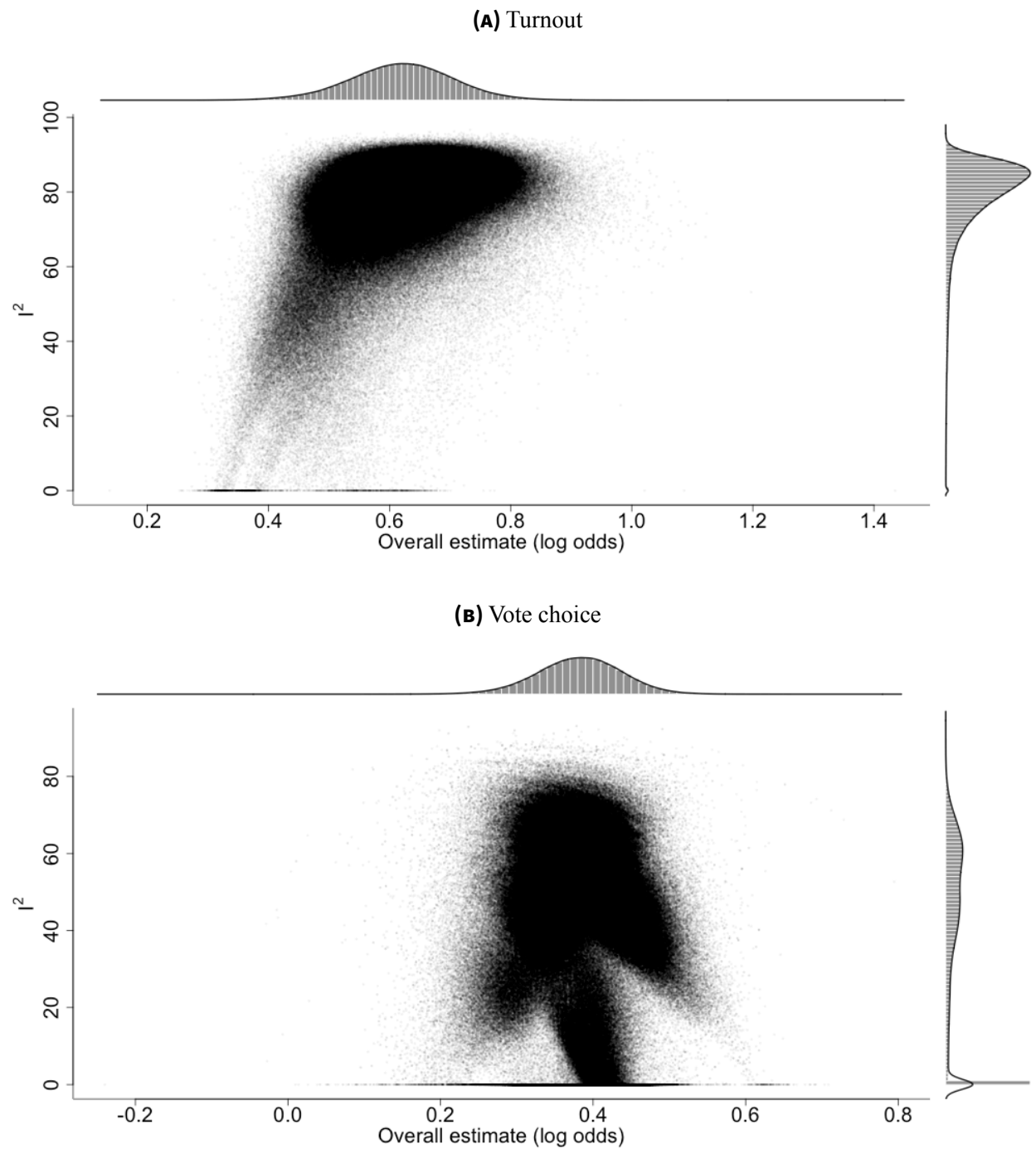


FIGURE B8: Sensitivity of overall effect estimates across several model specifications, by outcome



Note: The figure reports overall effect estimates together with 95% CIs across different model specifications by outcome type. The specifications used are (a) cross-classified random effects models (CCREM, as reported in the main text), (b) cross-classified random effects models excluding quasi duplicate studies that re-analyze study data which had been used in predecessor studies (CCREM, no dups), (c) fixed effects models (FE), standard random effects models (RE), random effects models assigning equal weights for effects (RE, equal weights), and random effects models assigning equal weights for studies (RE, study weights).

FIGURE B9: Graphical display of heterogeneity (GOSH) plot based on fixed-effects models in all possible subsets of effects.

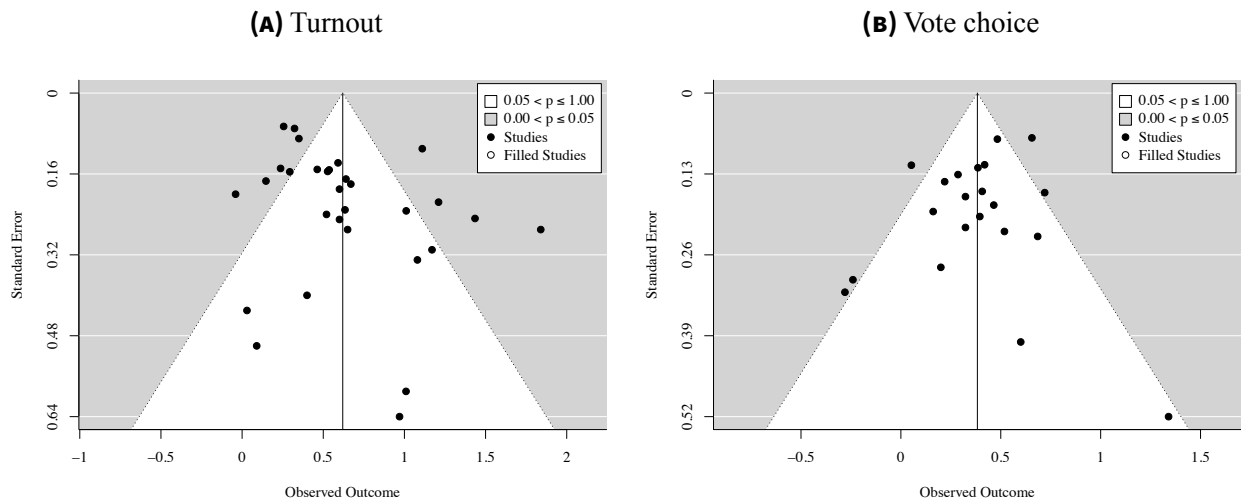


Note: The GOSH plot (Olkin et al., 2012) is based on fixed-effects models computed with all 2^{k-1} possible subsets of effects. The overall estimate is plotted against between-study heterogeneity I^2 .

APPENDIX C PUBLICATION BIAS ANALYSIS

To assess publication bias, we have utilized a trim-and-fill method, which takes a two-step approach to identifying and adjusting for publication bias based on the funnel plot (see [Duval and Tweedie, 2000](#)). First, the model trims out small- N studies to obtain a more symmetrical funnel plot and estimates a new summary effect based on the larger- N studies. Second, the model restores the trimmed studies and adds the imputed “missing” counterparts of the effects around the new summary effect estimate. There are no “missing” studies imputed in either case (see [Figure C1](#)). The results do not suggest support for potential publication bias for the reported effects. Given that the funnel plot-derived trim-and-fill method assumes publication bias as the only reason for asymmetry, we note that there are various alternative explanations in this case for the observed asymmetry, such as methodological heterogeneity.

FIGURE C1: Trim-and-fill funnel plots, random-effects models



APPENDIX D SOFTWARE STATEMENT

The entire analysis was run under OS X 10.15.4 using R version 3.6.2 (R Core Team, 2019). In the empirical analysis, we made use of the following R software packages:

dmetar (Harrer et al., 2019),
dplyr (Wickham and Francois, 2015),
ggplot2 (Wickham, 2016),
haven (Wickham and Miller, 2018),
janitor (Firke, 2018),
magrittr (Bache and Wickham, 2014),
metafor (Viechtbauer, 2010),
readxl (Wickham and Bryan, 2018),
stringr (Wickham, 2015),
writexl (Ooms, 2018), and
xtable (Dahl, 2016).

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